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# Understanding Consumer Decisions Using Behavioural Economics

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**Abstract:** Consumers make many decisions in everyday life involving finances, food and health. It is known from behavioural economics research that people are often driven by short-term gratification, i.e. people tend to choose the immediate, albeit smaller reward. But choosing the delayed reward, i.e. delaying the gratification, can actually be beneficial. How can we motivate consumers to resist the 'now' and invest in their future, leading to sustainable or healthy habits? We review recent developments from behavioural and neuroimaging studies that are relevant for understanding consumer decisions. Further, we present results from our field research that examined whether we can increase the perceived value of a (delayed) environmental benefit using tailored communication, i.e. change the way it is *framed*. More specifically, we investigated whether we can boost the value of an *abstract, long-term* 'green' claim of a product by expressing it as a *concrete, short-term* benefit. This is a new application area for behavioural economics.

**Key words:** Neuroeconomics; decision making; consumer preference; environmental benefits; health benefits; immediate rewards; delayed rewards; intertemporal choice.

## Introduction

Many decisions in day-to-day life involve considering alternatives that will occur at some point in the future. Consider for example, when given a choice between choosing \$5 now or \$10 in two weeks time, people tend to choose the immediate, albeit smaller reward. Recent developments in behavioural economics and neuroeconomics research has increased our understanding of how people make decisions in various contexts. Behavioural economics is a research field that uses psychological science to understand how people make economic decisions (Kahneman & Tversky, 1979; Camerer, 1999). This field recently expanded into neuroeconomics that looks at the role of the brain when people evaluate decisions, categorize risks and rewards, and how these interact with each other (Loewenstein *et al.*, 2008). Of particular interest is the time-discounted utility theory in this – so far mainly used in economics. Time-discounting describes the subjective devaluation of outcomes as a function of the time delay until their delivery, such that immediate rewards are more highly valued and hold a greater control over behaviour than those which are delayed (Frederick *et al.*, 2002). This chapter will discuss recent developments in our understanding of factors that influence consumers' decisions, with a particular focus on discounting behaviour around immediate and delayed rewards of fast moving consumer goods (FMCG). This is followed by some data on

42 boosting the value of delayed rewards in order to make them more motivating for consumers.  
43 Finally, we discuss the implications of such research and identify new avenues for research.

44 Psychological theories have emphasised the importance of rewards in decision making. Rewards  
45 have been defined operationally as stimuli which positively reinforce behaviour, and increase the  
46 frequency and intensity with which behaviours are elicited (Berridge & Robinson, 2003; McClure *et al.*, 2004b). Rewards may be either primary or secondary in nature. Primary rewards such as food,  
47 water or sexual stimuli, have the ability to reinforce behaviour without the need to be learnt,  
48 whereas secondary rewards, such as money, receive their reward value by being associated with  
49 primary rewards through conditioning (Rolls, 2007). In decision making situations involving choice  
50 behaviour, individuals choose the option that yields the highest expected reward value. The  
51 equivalent concept from behavioural economics is that of the utility function, which combines  
52 several parameters of each option into one single unitary value. In this way, decisions among  
53 several alternatives are driven by the aim of maximising utility. The key assumption here is that the  
54 utility function provides a way of quantifying the subjective reward value that an individual attaches  
55 to each option. We propose that the concept of utility functions can be applied to better understand  
56 consumer decision making in a principled way.  
57

58 In recent years, a lot of interest has arisen towards neuroeconomics to determine where aspects of  
59 decision making occur in the brain and to investigate how the brain computes reward value. Also,  
60 building on the utility framework, more and more studies addressed the question whether there  
61 exists a region of the brain that commonly encodes decision values for different types of rewards,  
62 or if, in contrast, the values of different types of rewards are represented in distinct brain regions  
63 (Chib *et al.*, 2009). The results of these studies have systematically revealed that the utility is  
64 represented in the so-called reward centers of the brain (Knutson *et al.*, 2005). Neurophysiological  
65 studies in animals have provided the primary basis for identifying the core neuronal circuitry  
66 underlying rewards. This core reward circuitry is composed of the ventral striatum and orbitofrontal  
67 cortex, both receiving inputs from the dopaminergic neurons in the substantia nigra and ventral  
68 tegmental areas in the midbrain (see Schultz (2000) for review). Neuroimaging studies in humans  
69 show similar findings and have revealed activation in analogous brain regions to various rewarding  
70 stimuli and predictors of reward (see O'Doherty (2004) for review). Chib *et al.* (2009) asked  
71 participants in the fMRI scanner to make purchasing decisions among different types of rewards  
72 (i.e. food, nonfood consumables and monetary gambles) and they found activity in the  
73 ventromedial frontal cortex to be correlated with subjects' valuations for all types of rewards. These  
74 results have been confirmed by Kim *et al.* (2011) who found that secondary rewards such as  
75 money activated the same reward regions as primary rewards such as juice (Kim *et al.*, 2011).  
76 Furthermore, pictures of appetizing foods activate the same reward regions of the brain as the  
77 actual experiences that occur while actually consuming these foods (Simmons *et al.*, 2005). These  
78 neural findings have significantly contributed to our understanding that a common reward valuation  
79 system is responsible for processing different types of reward.

80 Neuroeconomic methods have also shown that the neural mechanisms underlying consumer  
81 decisions could reliably predict consumer choices (see Kenning & Plassmann (2008) for a review;  
82 Tusche *et al.*, 2010). For example, Knutson *et al.* (2007) showed that nucleus accumbens activity

predicted an upcoming purchase decision when viewing a product, *even before price information was known*. In addition, Levy *et al.* (2011) showed that the activation in the striatum and medial prefrontal cortex during passive viewing of products reliably predicted subsequent choices made outside the scanner. In a similar vein, Plassmann *et al.* (2007) found that the medial orbitofrontal cortex encoded 'willingness to pay' for 50 different sweet and salty snack foods (e.g. chips and candy bars) by hungry subjects. Until now, as far as we are aware, most research has focused on preference for immediate rewards, and have not addressed how we can boost the utility of delayed rewards. However, several choices also involve making decisions regarding potential future rewards, a topic of high relevance to understand consumer decision behaviour.

## Intertemporal Choice & Delayed Gratification

Why is it difficult to save for a pension or maintain a healthy diet? Intertemporal choices are decisions about future and delayed consequences and occur in most of our decisions in daily life, such as adopting and maintaining a healthy lifestyle, investing in a health insurance or saving for a pension (Berns *et al.*, 2007). This temporal aspect in decision making involves trade-offs between immediate consumption and considerations about better future payoffs. Such intertemporal choices present a challenge for a decision maker, as there is a tendency to favour immediate gratification over benefits which are delayed in the future: instant gratification often appears to be far too much a temptation (Blackburn, 2012; Blackburn *et al.*, 2012). The discounted utility model (Samuelson, 1937) posits that decisions are made as a weighted sum of utilities with temporal discount factors as weights. This model assumes a constant discount rate, which ensures that a preference between two delayed options is only dependent on the time delay between them. The order of preferences should be preserved at all possible time points (time consistency or stationarity). This behaviour is implemented by an exponential form of discounting into the equation.

$$V = Ae^{-kd}, \text{ where } V \text{ is the discounted value, } A \text{ is the absolute value, } k \text{ is a constant discounting parameter and } d \text{ is the delay. (Equation 1)}$$

Consider the example when given a choice between choosing \$5 now or \$10 in two weeks. The smaller sooner reward (\$5 now) may be preferred compared to the larger delayed reward (\$10 in 2 weeks) because the value of the future reward is discounted below the value of the immediate reward (Equation 1; Figure 1). Now, consider that both options in the above gamble are further delayed by one week. The exponential model of discounting suggests that the smaller sooner reward (\$5 in one week) would still be preferred over the larger later reward (\$10 in 3 weeks). However, behavioural experiments showed that people fail to conform to such a prediction (Berns *et al.*, 2007). Thus, a preference reversal occurs if people prefer the larger delayed reward in one but not the other situation. This behaviour is captured by the hyperbolic discounting model (Equation 2; Figure 1). The latter finding is particularly relevant for consumer behaviour, i.e. how to motivate consumers to resist the now and invest in future rewards.

$$V = \frac{A}{1+kd}, \text{ where the different symbols hold the same meaning as Equation 1. (Equation 2)}$$

-- INSERT FIGURE 1 HERE--

Neuroeconomic findings argue that immediate and future rewards are represented by either a unitary system, in which discounting arises as a function of a single valuation mechanism, or a dual system, which focuses on the interaction between two separate decision processes that have competing goals. Neuroeconomic research tried to address both discounting mechanisms in decision making. For example, Kable and Glimcher (2007) showed that neural activity in ventral striatum, medial prefrontal cortex and posterior cingulate cortex tracked the subjective value of delayed monetary rewards when participants choose between a fixed immediate reward of \$20 and a larger delayed reward, a value in the range from \$20.25 to \$110, available after a delay ranging from 6 hours to 180 days. With the help of a quasi-hyperbolic discount function, McClure *et al.* (2004a) found separate neural systems underlying immediate and delayed rewards. They examined neural correlates of time discounting while subjects made a series of choices between monetary reward options that varied by delay (same day to 6 weeks later) to delivery. They found that ventral striatum, medial OFC, medial prefrontal cortex (PFC), posterior cingulate and left posterior hippocampus were related to choices of immediate rewards. In contrast, regions of lateral prefrontal cortex and posterior parietal cortex were engaged uniformly by intertemporal choices irrespective of delay. In a subsequent study, McClure *et al.* (2007) used primary rewards (fruit juice or water) with time delays of minutes instead of weeks and found similar activation patterns as in their previous study. The studies of Kable & Glimcher (2007) and McClure *et al.* (2004a; 2007) provided evidence for unitary and dual system accounts of processing immediate and delayed rewards, respectively.

Prior research has mainly considered temporal discounting of monetary rewards in humans and primary rewards such as food in animals. Given that the neural substrates of different goods overlap, it is critical to consider whether temporal discounting behavior is domain specific or not. Tsukayama and Duckworth (2010) examined intertemporal choice for money (dollars) and three food-related consumables (candy bars, chips and beer) at five different delays – one week, one month, six months, one year, and three years. A domain-general discounting was revealed by positive correlation for discounting rates for the four categories. However, steeper discounting was found for items that individuals found more tempting. In another study comparing financial, environmental and health gains or losses available immediately or after a delay of 1 or 10 years, Hardisty and Weber (2009) found that discount rates are influenced more by the valence of outcomes (gains or losses) than by domain (i.e. money, environment or health). More importantly, at short or medium delays, they found that environmental outcomes were discounted in a similar way as financial outcomes. This recent behavioural work opens a new area of application for behavioural economics to outcomes involving health and environmental benefits.

An active area of research parallel to intertemporal choice paradigms is that of delayed gratification – the ability to wait over a delayed time interval to receive the desired reward. In the classic marshmallow test by Mischel *et al.* (1989), pre-school children were encouraged to resist the temptation of eating a single piece of their favourite confectionary while the experimenter was absent from the room up to 15 minutes. In exchange, children who could successfully delay the gratification would be offered two pieces of their preferred confectionary. The observation of

163 childrens' behavior revealed different cognitive strategies - by focusing on the 'hot' properties such  
164 as the taste of the confectionary led the child to be more impulsive, while focusing on the 'cool'  
165 properties such as the shape of the object or other types of distractions, children were able to wait  
166 for longer. A remarkable finding is that the delay gratification measures of these pre-schoolers  
167 could successfully predict their scholastic achievements as adolescents (Mischel *et al.*, 1989). A  
168 recent brain imaging study examined cognitive control strategies on a subset of the original study's  
169 participants using a go/no-go paradigm (Casey *et al.*, 2011). Activation in the inferior frontal gyrus  
170 was greater for high delayers, whereas activation in ventral striatum was greater for low delayers.  
171 The findings from these longitudinal studies (spread over four decades) expand our knowledge on  
172 individual differences in self-control and the appropriate strategies for cognitive control.

### 173 **Cognitive Modulation of Immediate Rewards**

174 Reward processing has received a lot of attention in psychology and in the neurosciences, and our  
175 understanding of neurological systems underlying reward processing has increased tremendously  
176 in the last decade. To date, most of this research has focused on bottom-up processing, i.e.  
177 involving stimuli that are innately seen as rewards/punishments. Less attention has, however, been  
178 paid to top-down processing, i.e. the influence of expectations – based on prior knowledge, our  
179 own thoughts and ideas, information, emotions – on the processing of rewards.

180 Expectations are an important tool for humans, helping us to predict daily life and to deal with  
181 uncertainties. These expectations have an influence on how we perceive and even on what we  
182 perceive. For consumers, expectations of product performance through price or brand  
183 communication such as advertising and packaging play a significant role in deciding whether to  
184 buy a product or not. There are a number of behavioural food studies that showed that  
185 expectations influence the perception and reward value of a food (Cardello, 2007; Deliza &  
186 MacFie, 1996; Winkielman *et al.*, 2005). For example, the expectation of a red wine will change the  
187 taste description of a coloured white wine (Morret *et al.*, 2001), and believing in the effect of an  
188 ingredient can lead to a strong placebo effect (Kuenzel *et al.*, 2011).

189 More and more neuroimaging studies have been conducted that support that expectations  
190 influence the reward value of a stimulus (de Araujo *et al.*, 2005; McClure *et al.*, 2004b; Plassman *et al.*,  
191 2008). A key experiment was carried out by de Araujo *et al.* (2005) using the odour isovaleric  
192 acid. They presented participants this odour in combination with different cognitive labels. When  
193 the odour was labelled with 'cheese', subjects expressed liking the odour more than when it was  
194 labelled with 'body odour'. Moreover, different areas in the brain were activated depending on the  
195 cognitive labels presented. Another study focussed on the effect of brand preference on reward  
196 processing of a primary reward like a sugared drink (McClure *et al.*, 2004b). McClure *et al.* (2004b)  
197 presented participants two drinks, Pepsi and Coke, which are essentially sweet brown liquids with  
198 similar taste and sugar content. They gave participants the drinks unbranded and showed that the  
199 sugar in it leads to an activation of reward processes in the brain, more specifically in the  
200 ventromedial prefrontal cortex. Interestingly, they then gave participants the drinks branded and  
201 showed that -for subjects who preferred Coke over Pepsi- more reward-related brain areas were  
202 activated when the Coke brand was shown, compared to when the Pepsi brand was shown. These

203 brain areas (the hippocampus and the dorsolateral prefrontal cortex) have been associated with  
204 emotion and affect. This shows that the reward can be generalised to a brand. In a similar vein,  
205 Plassmann *et al.* (2008) showed that other marketing actions, such as changes in price of a  
206 product, can affect neural representations of experienced pleasantness. They asked subjects to  
207 taste (identical) wines that they believed to be different and sold at different prices. They  
208 hypothesised that a higher price would result in higher taste expectations and hence would lead to  
209 higher activity in the medial orbitofrontal cortex (mOFC), an area in the brain that is widely thought  
210 to encode for experienced pleasantness. The results were consistent with this hypothesis: the  
211 reported price of wines markedly affected self-reported experienced pleasantness and, more  
212 importantly, also modulated the blood-oxygen-level-dependent (BOLD) signal in the mOFC. These  
213 results show that a reward value of a food not only depends on the sensory properties of the food  
214 and/or the current state of the consumer: the brain computes a reward value in a much more  
215 sophisticated manner that involves integrating the sensory properties of the food being consumed  
216 with the expectations about how good it should be.

217 Little is known about how top-down effects influences choice behaviour. In a laboratory behavioural  
218 experiment using a within-subjects design, Zandstra & El-Deredy (2011) gave 44 subjects one of  
219 two yoghurt drinks, alternating between high-energy and low-energy yoghurt drinks (255 kcal and  
220 57 kcal per 200ml serving, respectively), first thing in the morning following 8 hours of fasting,  
221 every day for two weeks, with 5 exposures to each yoghurt drink on alternate days. Both drinks  
222 were similar in appearance, texture, taste etc. Participants were not aware that the drinks differed  
223 in any way, except that one drink was paired with a blue label, and the other with a pink label, with  
224 the pairings fully counter-balanced. Every day of the third (test) week, participants were given the  
225 choice of either having the pink or the blue labeled yoghurt drink. Participants did not differentiate  
226 between the yoghurt drinks in terms of how much they said they liked them. However, clear and  
227 pronounced effects were found for the behaviour measure preference: even though the yoghurt  
228 drinks were *equally* liked, participants chose the high energy drink significantly more often over the  
229 low energy drink. This suggests a conditioned preference for a delayed physiological (energy)  
230 reward. Further evaluative conditioning research with food stimuli under realistic choice conditions  
231 in a natural (everyday life) context is needed to confirm this conclusion. We are now beginning to  
232 study the neural correlates of this behaviour with a view to being able to better understand and  
233 perhaps predict subjects' choices. Are different reward areas of the brain being activated by the  
234 two drinks? Is there any mismatch between what we expect and what we observe?

235 To recapitulate, just as in behavioural studies, neuroimaging studies do support the influence of a  
236 number of different top-down effects on the reward value of stimuli. These top-down influences are  
237 strong enough to override bottom-up effects such as sensory stimulation and they potentially drive  
238 choice behaviour. Therefore, it is critical to take all these different aspects of a product into account  
239 to create a maximised *immediate* rewarding experience.

## 240 **Boosting Delayed Rewards**

241 An immediate rewarding experience is delivered directly at the point of consumption (e.g. price,  
242 efficacy), whereas delayed rewards are characterized by a delay, i.e. it takes months or even years

243 until effects can be measured (e.g. health or sustainability). Even though delayed rewards are  
244 motivating for consumers in the long-term, it takes time before they are actually perceived by  
245 consumers. Decision making in relation to health and sustainability is therefore not readily  
246 predictable as these rewards are less tangible, more impersonal and delayed. Consumers will ask  
247 themselves: 1) “will I receive a reward?”, 2) “when will I receive a reward?”, and 3) “what reward  
248 will I receive?”. As mentioned before, behavioural economics research shows that consumers are  
249 often driven by short-term gratification: people tend to choose the sooner, smaller reward. It is  
250 therefore critical to address this for behaviours around health and sustainability. The challenge  
251 here will be to turn delayed and impersonal rewards into personally (immediately) perceivable  
252 rewards. This, in the end, will maximise the *delayed* rewarding experience.

253 So far, the paradigms that tested for choices between immediate and delayed rewards have mainly  
254 been concerned with impulsive choices (e.g. McClure *et al.*, 2004a). Little is known about how to  
255 best promote choices of products with delayed rewards. When it comes to higher order choice  
256 motivations such as health and sustainability, there is a considerable attitude-behaviour gap  
257 (Sheeran, 2002), i.e. consumers “talk healthy” or “talk green” but “don’t walk it”. For example,  
258 Weijzen *et al.* (2008) found a substantial inconsistency between healthful stated snack choice  
259 intentions and actual behaviour. Participants were asked about their intentions in choosing among  
260 four snacks: an apple, a banana, a candy bar and a waffle. About half of the participants indicated  
261 they would choose the apple or banana - a “healthy” snack. But when presented with the actual  
262 snacks one week later, 27% switched to the candy bar or waffle. Over 90% of the unhealthy-choice  
263 participants stuck with their intentions and chose the unhealthy snack. This demonstrates that  
264 intentions are usually under cognitive control while actual choices are often made impulsively, even  
265 unconsciously.

266 Behavioural economics might enable us to explore the dynamics of these decisions in more detail.  
267 The behavioural economics approach has been investigated extensively in the financial domain.  
268 Recent developments show that this approach can be extended to other domains such as food  
269 (Epstein *et al.*, 2010). Consider the following scenario: given a choice between a tasty, but  
270 potentially unhealthy food (e.g. a burger) available immediately and a second option which involves  
271 resisting the temptation now and choosing to have a healthy weight in future, usually people would  
272 want both. However, the consequences and antecedents for these two choices are very different.  
273 Choosing a high fat food is ultimately detrimental to your body weight and the choice of maintaining  
274 a healthy weight requires eating low fat foods such as vegetables. In order to promote the healthy  
275 option, we could potentially think of two policies – either make people pay more for the burger or  
276 make vegetables a much cheaper food. Classical economics theory suggests that if the price of a  
277 particular food *decreases*, choice for that food will *increase*, and vice versa. However, recent  
278 research by Epstein *et al.* (2010) showed that providing subsidy on healthier foods is not a good  
279 strategy, as the consumers will use the money saved to invest back into the unhealthy option.

280 In our current research, we explored whether the behavioural economics approach can be  
281 extrapolated to sustainability as well. As indicated before, environmental benefits are extremely  
282 delayed in time, which makes them less tangible compared to immediate benefits, such as price  
283 and efficacy. A consumer is unlikely to directly experience many environmental benefits in his/her



284 life time. A green claim, therefore, is not a directly perceivable reward and appeals to consumers  
285 only through their individual belief in environmental benefits (Bolderdijk, 2011). Moreover, in our  
286 research we draw on construal level theory and build on the assumption that environmental  
287 benefits are not the primary, but are rather secondary drivers of product choice because of their  
288 perceived psychological distance to the 'now', *i.e.* a primary benefit 'what the product does for you'  
289 vs. a secondary benefit 'with what social and ecological impact the product does it' (Trope *et al.*,  
290 2007; Trope & Liberman, 2010). Both primary and secondary benefits may be salient and relevant  
291 to the consumer, but in terms of determinance in guiding behaviour primary benefits typically take  
292 priority. Especially when primary and secondary benefits are to some extent at odds, social  
293 dilemma theory indeed indicates that it is more likely that primary benefits will be pursued to direct  
294 need satisfaction than that people sacrifice their immediate self-interest for long-term collective  
295 benefit (Balliet *et al.*, 2011; Van Lange *et al.*, 1998).

296 Therefore, the question we asked ourselves is whether we can boost the utility curve of a delayed  
297 reward in a product by re-framing it as a primary benefit. For example, one could frame a green  
298 claim by only emphasising the green benefit (e.g. 'With this soup we help to protect the planet's  
299 natural resources, as it is made with eco-friendly tomatoes.'), or one could frame a green claim by  
300 expressing it as a primary benefit (e.g. 'This soup is a great tasting soup, as it is made with eco-  
301 friendly tomatoes, which also helps to protect the planet's natural resources.').

302 Previous research showed that re-framing delayed outcomes can impact the ability to consider  
303 future consequences (Tversky & Kahneman, 1981). One approach is to raise awareness about  
304 environmental issues and the individual consumer role and responsibility by educating the  
305 consumer using explicit and tailored information, spelling out what the consumer needs to do. For  
306 example, in hotel rooms guests are generally pointed at the threat posed to the environment from  
307 the daily cleaning of towels. They are kindly asked to re-use their towels in order to protect the  
308 environment by not dropping them on the floor. Goldstein *et al.* (2008) demonstrated that when  
309 guests were made aware of the fact that "the majority of guests *in this room* have re-used their  
310 towels", they were more likely to re-use their towels as well. Interestingly, just by re-framing it and  
311 making it more personally relevant, consumers showed more sustainable behaviour.

## 312 **Experiment into boosting delayed rewards**

313 In a laboratory experiment, we examined whether we can increase the perceived value of a  
314 (delayed) environmental benefit using tailored communication, *i.e.* by changing the way it is  
315 *framed*. More specifically, we investigated whether we can boost the value of an *abstract, long-*  
316 *term* green claim in a home care product by expressing it as a *concrete, short-term* benefit. The  
317 key question was: Which combinations of primary and green benefits works the best in boosting  
318 the utility of the delayed reward?

319 *Participants* - A total of 254 consumers were recruited within the UK for an in-home or at work  
320 internet-based study. Among these were 117 males and 137 females with an average age of  $39 \pm$   
321 9.6 years. The sample was representative of the wider population in terms of socioeconomic  
322 status, educational level and income level. Participants received a small incentive for participation  
323 in the study.

324 *Stimuli* - We tested six primary benefits with four green claims. The green claims varied on three  
325 parameters: type of green claim, quantity of green claim and time to achieve the green claim. From  
326 available data from an environmental calculator at internet, average savings per months were  
327 calculated for each claim type (e.g. saving 500 trees in 1 month). Four time points (3 months, 6  
328 months, 1 year and 2 years) were chosen to enable us to construct the utility curves. To match with  
329 the four time points, the average monthly environmental savings were multiplied by the duration  
330 (i.e. 3, 6, 12, or 24 months) to give four different quantities. For a given green claim type, the  
331 quantity and time were varied independently, with the constraint that any given choice set with a  
332 repeated type of green claim should satisfy the following condition: longer time to achieve the  
333 green claim should always be accompanied by increased quantity of environmental savings. A  
334 linear relationship between time and quantity was not enforced to ensure that these two  
335 parameters were orthogonal. The research reported here focusses on the results of only one of the  
336 four green claims as similar results were observed for the other green claims.

337 *Procedure* - Using a conjoint, 3-alternative forced choice approach, we asked consumers to  
338 choose one among three pairs of statements: "Which statement would motivate you the most to  
339 buy home care product 'X'?" Each pair of statements consisted of a primary and a green claim.  
340 Participants indicated their choice by selecting a radio button at the bottom of the screen. In total,  
341 participants completed 8 such choice tasks.

342 *Data analysis* - Hierarchical Bayesian models were fitted to the observed choice data. The choice  
343 of models was based on their root likelihood ratio (RLH). The maximum RLH was set to 1000. If  
344 respondents chose completely randomly among the three alternatives that were presented, the  
345 root likelihood ratio (RLH) would be  $1000/3=333$ . Three Hierarchical Bayesian models were fitted  
346 to the choice data: model I included only the main effects (RLH 535), model II included main  
347 effects and significant interaction effect of (green claim)x(quantity of green claim) (RLH 557), and  
348 model III included main effects with relevant interaction effects, i.e. (primary claim)x(green claim),  
349 (type of green claim)x(quantity of green claim), (type of green claim)x(time to achieve green claim),  
350 (quantity of green claim)x(time to achieve green claim) (RLH 659). Model III with the highest RLH  
351 was used for reporting the results. The advantage of using a hierarchical bayesian model over  
352 traditional models such as logit was that we could estimate the utilities for each participant for *all*  
353 the choices, even though they would have actually responded to only a subset of the choice set.

354 -- INSERT FIGURE 2 HERE--

355 *Results* - Figure 2 shows the perceived value of one of the green claims using tailored  
356 communication. Inspection of Figure 2 shows that the perceived value (estimated utilities)  
357 decreased over time corresponding to standard utility curves in behavioural economics. Also, we  
358 observed a shift in the utility by expressing the green claim as a primary benefit. Statistical analysis  
359 confirmed these clear differences in utility curves ( $p<0.05$ ), even though the discounting factor ( $k$ )  
360 was small. Further, we do not notice a change in the slope of the utility curves [no significant  
361 interaction effect ( $p>0.05$ ) of (green claim)x(primary benefit)].

362 *Conclusion* - We confirmed our hypothesis that by appropriately communicating the green claim as  
363 a primary benefit, we can boost the perceived value of it. Therefore, to motivate consumers to buy

364 a product with an environmental benefit, it needs to have a green claim expressed as a concrete  
365 and short-term benefit.

## 366 **Conclusions and Future Directions**

367 Sustainability is a new application area for behavioural economics. This approach for the study of  
368 delayed rewards can be extended to other domains such as reducing obesity (Just & Payne, 2009)  
369 and improving cardiovascular health (Fair *et al.*, 2008). The research in this area will probably be  
370 most successful if it takes a holistic approach into account, i.e. focusing on product, packaging and  
371 communication as well as on the consumer and the context in which a product is consumed and  
372 used.

373 Future research should focus on the neural integration of primary and environmental benefits. So  
374 far, it is unclear whether environmental benefits are processed in the same way as primary benefits  
375 and whether the two are actually incorporated into a single utility processing module in the brain. If  
376 this were the case, then trade-offs between immediate primary and delayed environmental benefits  
377 would indeed be real and one would be more confident that one can substitute for the other. In a  
378 similar vein, it would be of interest to investigate whether environmental benefits are encoded in  
379 the activations of brain regions that process also primary benefits and how these or other brain  
380 activations predict actual choice behaviour. For the latter, it is critical to design studies to better link  
381 brain responses to real decisions, and there is a clear need for long-term studies of repeated  
382 choice and consumption.

383 Over the past decade we have learned that there are tremendous individual differences in brain  
384 responses. We therefore expect that there will be individual differences in the degree to which  
385 primary and environmental benefits are weighted. A brain region integrating the two types of  
386 benefits would show stronger activation to higher levels of primary benefits and further  
387 enhancement of signals by larger environmental benefits if participants value these benefits  
388 positively or suppression of signal if they value them negatively. Moreover, activation should be  
389 equal if the sum of environmental and primary benefits is equal, such that an increase in one can  
390 substitute for a decrease in the other. We predict such brain regions to include the striatum and  
391 parts of the prefrontal cortex.

392 To conclude, in this chapter we have explored the behavioural economics approach with a focus  
393 on delayed rewards. Behavioural economics studies have provided a principled approach to  
394 understand consumer decisions. Neuroeconomics research gives us three unique insights in  
395 understanding consumer decisions: 1) the core reward circuitry in the brain is activated by different  
396 kinds of rewards, 2) temporal aspects of decisions, i.e. immediate and delayed rewards, are  
397 represented in the brain, and 3) expectations influence the perceived value of rewards. Together,  
398 behavioural and neuroimaging findings have suggested that 'framing' could be a way forward to  
399 boost the value of delayed rewards. We confirmed our hypothesis that by appropriately  
400 communicating a green claim, we can boost the perceived value of it. Therefore, to motivate  
401 consumers to buy a product with an environmental benefit, it needs to have a green claim  
402 expressed as a concrete and short-term benefit. This is a new application area for behavioural  
403 economics, which has so far been mainly implemented for monetary rewards. There is much more  
404 potential for consumer relevant research, for example in rational decision making and prediction of  
405 actual choice behaviour.

406

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410 **References**

- 411 de Araujo, I.E., Rolls, E.T., Velazco, M.I., Margot, C. & Cayeux, I. (2005). Cognitive modulation of  
412 olfactory processing. *Neuron*, 46: 671-679.
- 413 Balliet, D., Mulder, L.B., & Van Lange, P.A.M. (2011) Reward, punishment, and cooperation: A  
414 meta-analysis. *Psychological Association*, 137(4): 594-615.
- 415 Berns, G.S., Laibson, D., & Loewenstein, G. (2007). Intertemporal choice - Toward an integrative  
416 framework. *Trends in Cognitive Sciences*, 11: 482-488.
- 417 Berridge, K.C. & Robinson, T.E. (2003). "Parsing reward." *Trends in Neurosciences*, 26(9): 507-  
418 513.
- 419 Blackburn, M. (2012). Decision making under uncertainty: Differentiating between 'If', 'What' and  
420 'When' outcomes occur. *PhD Thesis*, University of Manchester, U.K.
- 421 Blackburn, M., Liam, M., Hoeksma, M., Zandstra, E.H. & El-Deredy, W. (2012). Delay discounting  
422 as emotional processing: An electrophysiological study. *Cognition & Emotion*, iFirst article,  
423 1-16, DOI:10.1080/02699931.2012.673478.
- 424 Bolderdijk, J.W. (2011). Buying people: The persuasive power of money. *PhD Thesis*  
425 Rijksuniversiteit Groningen, The Netherlands. <http://irs.ub.rug.nl/ppn/334141206>.
- 426 Camerer, C. (1999). Behavioral economics: Reunifying psychology and economics. *Proceedings of*  
427 *the National Academy of Sciences of the U.S.A.* 96(19): 10575-10577.
- 428 Cardello, A.V. (2007). Measuring consumer expectations to improve food product development. In  
429 H.J.H. Macfie (Ed.) *Consumer-led food product development*. Cambridge: Woodhead  
430 Publishing pp. 223-261.
- 431 Casey, B.J., Somerville, L.H., Gotlib, I.H., Ayduk, O., Franklin, N.T., Askren, M.K., Jonides, J.,  
432 Berman, M.G., Wilson, N.L., Teslovich, T., Glover, G., Zayas, V., Mischel, W. & Shoda, Y.  
433 (2011). Behavioral and neural correlates of delay of gratification 40 years later. *Proceedings*  
434 *of the National Academy of Sciences of the U.S.A.*, 108(36): 14998-5003.
- 435 Chib, V.S., Rangel, A., Shimojo, S. & O'Doherty, J.P. (2009). Evidence for a common  
436 representation of decision values for dissimilar goods in human ventromedial prefrontal  
437 cortex. *Journal of Neuroscience*, 29(39): 12315-12320.
- 438 Deliza, R. & Macfie, H.J.H. (1996). The generation of sensory expectation by external cues and its  
439 effect on sensory perception and hedonic ratings: A review. *Journal of Sensory Studies*, 11:  
440 103-128.
- 441 Epstein, L.H., Dearing, K.K., Roba, L.G., & Finkelstein, E. (2010). The influence of taxes and  
442 subsidies on energy purchased in an experimental purchasing study. *Psychological*  
443 *Sciences*, 21(3):406-14.
- 444 Fair, A.K., Murray, P.G., Thomas, A. & Cobain, M.R. (2008). Using hypothetical data to assess the  
445 effect of numerical format and context on the perception of coronary heart disease risk.  
446 *American Journal of Health Promotion*, 22(4):291-6.

447 Frederick, S., Loewenstein, G. & O'Donoghue, T. (2002). Time discounting and time preference: A  
448 critical review. *Journal of Economic Literature*, 40: 351–401.

449 Goldstein, N.J., Cialdini, R.B. & Griskevicius, V. (2008). A room with a viewpoint: Using social  
450 norms to motivate environmental conservation in hotels. *Journal of Consumer Research*,  
451 35: 472-482.

452 Hardisty, D.J. & Weber, E.U. (2009). Discounting future green: money versus the environment.  
453 *Journal of Experimental Psychology: General*, 138(3): 329-40.

454 Just, D.R. & Payne, C.R. (2009). Obesity: can behavioral economics help? *Annals of Behavioural*  
455 *Medicine*, 38(1): S47-S55.

456 Kable, J.W. & Glimcher, P.W. (2007). The neural correlates of subjective value during  
457 intertemporal choice. *Nature Neuroscience*, 10(12): 1625-1633.

458 Kahneman, D. & Tversky, A. (1979). Prospect Theory: An Analysis of Decision under Risk.  
459 *Econometrica*, 47(2), 263-291.

460 Kenning, P.H. & Plassmann, H. (2008). How neuroscience can inform consumer research. *IEEE*  
461 *Trans Neural Systems and Rehabilitation Engineering*, 16(6): 532-538.

462 Kim, H., Shimojo, S. & O'Doherty, J.P. (2011). Overlapping responses for the expectation of juice  
463 and money rewards in human ventromedial prefrontal cortex. *Cerebral Cortex*, 21(4): 769-  
464 776.

465 Knutson, B. & Peterson, R. (2005). Neurally reconstructing expected utility. *Games and Economic*  
466 *Behavior*, 52(2): 305-315.

467 Knutson, B., Rick, S., Wimmer, G.E., Prelec, D. & Loewenstein, G. (2007). Neural predictors of  
468 purchases. *Neuron*, 53(1): 147-56.

469 Kuenzel, J., Barton, C., Blanchette, I., Zandstra, E.H., Thomas, A., & El-Deredy, W. (2011).  
470 Awareness changes placebo effects for feeling relaxed, but not for liking. *Journal of*  
471 *Marketing Communications*, iFirst article: 1-18.

472 Levy, I., Lazzaro, S.C., Rutledge, R.B. & Glimcher, P.W. (2011). Choice from non-choice:  
473 predicting consumer preferences from blood oxygenation level-dependent signals obtained  
474 during passive viewing. *Journal of Neuroscience*, 31(1): 118-125.

475 Loewenstein, G., Rick, S. & Cohen, J.D. (2008). Neuroeconomics. *Annual Review of Psychology*,  
476 59: 647-672.

477 McClure, S.M., Laibson, D.I., Loewenstein, G. & Cohen, J.D. (2004a). Separate neural systems  
478 value immediate and delayed monetary rewards. *Science*, 306(5695): 503-507.

479 McClure, S.M., Li, J., Tomlin, D., Cypert, K.S., Montague, L.M. & Montague, P.R. (2004b) Neural  
480 Correlates of Behavioral Preference for Culturally Familiar Drinks. *Neuron*, 44: 379–387.

481 McClure, S.M., Ericson, K.M., Laibson, D.I., Loewenstein, G. & Cohen, J.D. (2007). Time  
482 discounting for primary rewards. *Journal of Neuroscience*, 27(21): 5796-5804.

483 Mischel, W., Shoda, Y. & Rodriguez, M.L. (1989). Delay of gratification in children. *Science*, 244:  
484 933-938.

485 Morrot, G., Brochet, F. & Dubourdieu, D. (2001). The color of odors. *Brain and Language*, 79, 309-  
486 320.

487 O'Doherty, J.P. (2004). Reward representations and reward-related learning in the human brain:  
488 Insights from neuroimaging. *Current Opinion in Neurobiology*, 14(6): 769-776.

489 Plassmann, H., O'Doherty, J. & Rangel, A. (2007). Orbitofrontal cortex encodes willingness to pay  
 490 in everyday economic transactions. *Journal of Neuroscience*, 27(37): 9984-9988.

491 Plassmann, H., O'Doherty, J., Shiv, B. & Rangel, A. (2008). Marketing actions can modulate neural  
 492 representations of experienced pleasantness. *Proceedings of the National Academy of*  
 493 *Sciences of the U.S.A.*, 105(3): 1050-1054.

494 Rolls, E.T., McCabe, C., & Redouter, J. (2007). Expressed value, reward outcome, and temporal  
 495 difference error representation in a probabilistic decision task. *Cerebral Cortex*, 18(3):652-  
 496 663.

497 Samuelson, P. (1937). A note on measurement of utility. *The Review of Economic Studies*, 4: 155-  
 498 161.

499 Simmons, W.K., Martin, A., & Barsalou, L.W. (2005). Pictures of appetizing foods activate  
 500 gustatory cortices for taste and reward. *Cerebral Cortex*, 15, 10:1602-8.

501 Schultz, W. (2000). Multiple reward signals in the brain. *Nature Reviews Neuroscience*, 1(3): 199-  
 502 207.

503 Sheeran, P. (2002). Intention-behaviour relations: A conceptual and empirical review. *European*  
 504 *Review of Social Psychology*, 12: 1-36.

505 Trope, Y., & Liberman, N. (2010). Construal-level theory of psychological distance. *Psychological*  
 506 *Review*, 117(2): 440-463.

507 Trope, Y., Liberman, N., & Wakslak, C. (2007). Construal levels and psychological distance:  
 508 Effects on representation, prediction, evaluation, and behaviour. *Journal of Consumer*  
 509 *Psychology*, 17(2): 83-95.

510 Tsukayama, E., & Duckworth, A.L. (2010). Domain-specific temporal discounting and temptation.  
 511 *Judgment and Decision Making*, 5(2): 72-82.

512 Tusche, A., Bode, S., & Haynes, J.D. (2010). Neural responses to unattended products predict  
 513 later consumer choices. *Journal of Neuroscience*, 30(23): 8024-8031.

514 Tversky, A., & Kahneman, D. (1981). The Framing of Decisions and the Psychology of Choice.  
 515 *Science*, 211: 453-458.

516 Van Lange, P.A.M., Van Vugt, M., Meertens, R.M., & Ruiter, R.A.C. (1998). A social dilemma  
 517 analysis of commuting preferences: The roles of social value orientation and trust. *Journal*  
 518 *of Applied Social Psychology*, 28(9): 796-820.

519 Weijzen, P.L.G., de Graaf, C., & Dijksterhuis, G.B. (2008). Discrepancy between snack choice  
 520 intentions and behavior. *Journal of Nutrition Education and Behavior*, 40: 311-316.

521 Winkielman, P., Berridge, K.C., & Wilbarger, J.L. (2005). Unconscious Affective Reactions to  
 522 Masked Happy Versus Angry Faces Influence Consumption Behavior and Judgments of  
 523 Value. *Personality and Social Psychology Bulletin*, 31(1): 121-135.

524 Zandstra, E.H. & El-Deredy, W. (2011). Effects of energy conditioning on food preferences and  
 525 choice. *Appetite*, 57: 45-49.

526

**FIGURES**

**Understanding      Consumer      Decisions      Using      Behavioural      Economics**  
EH Zandstra *et al.*

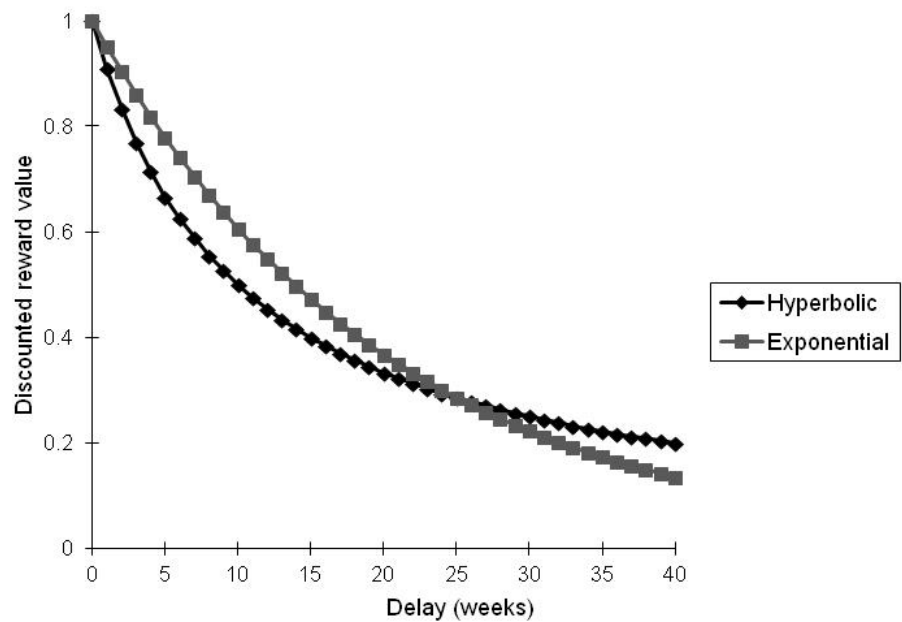


Figure 1. Example of discounted utility functions defined by exponential and hyperbolic function. Discounted values are calculated as per Eq.s 1 and 2, respectively.

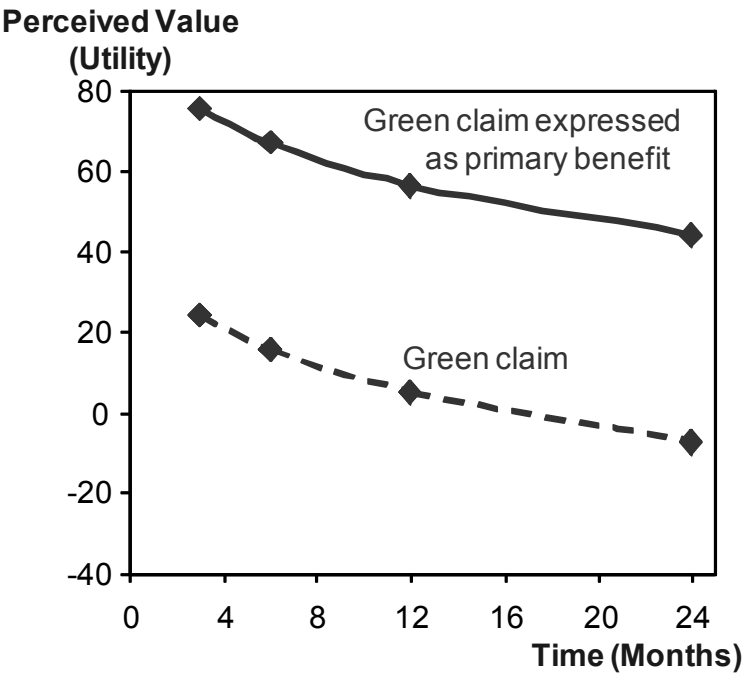


Figure 2. Boosting the perceived value of a delayed reward using tailored communication. The y-axis shows perceived values (estimated utilities), for one of the four green claims (dotted line) and the corresponding one of six primary benefits, which was found to be an optimal combination. The

*solid line is obtained by summing the utilities of the green claim, the primary benefit and interaction term. Each of the main estimated utility terms are zero-centred by definition. Hence, the sum of them need not always be higher than the individual terms. The relative value can be compared between any pair of points.*